Appendix C.
Statistical Methodology

THE CENSUS MAIL LIST AND SCREENER PHASE

The National Agricultural Statistics Service (NASS) maintains a list of farmers and ranchers from which the census mail list (CML) is compiled. The goal is to build as complete a list as possible of agricultural places that produce and sell, or would normally sell, $1,000 or more of agricultural products. This is the same list used to define sampling populations for NASS surveys conducted for the agricultural estimates program. Each record on the list includes name, address, and telephone number plus additional information used to efficiently sample and administer the NASS census of agriculture and its agricultural estimates programs.

NASS builds the list on an ongoing basis by obtaining outside source lists to improve the list sampling frame. Sources include state and federal government lists, producer a field office association lists, seed grower lists, pesticide applicator lists, veterinarian lists, marketing a field office association lists, and a variety of other agriculture related lists. NASS occasionally obtains special commodity lists to address specific list deficiencies. In 2000, NASS began an intensive push to increase list coverage in preparation for the census.

Most names on a newly acquired list are already on the list sampling frame. Those found on the list are set aside. Those not found are treated as potential farms until NASS can confirm their existence as a qualifying farm. State offices routinely contact these potential farms to determine their status, however, the increased pre-census list building activity generates much more follow up work.

Beginning in April 2002, NASS conducted the Farm Identification Survey to screen 591,288 potential farms before placing them on the CML. These records were mailed a short report form and a non-response follow up mailing was made in May 2002. A second group of 568,692 additional potential farm records was pulled in late June 2002. A single mailing was made to this group. The entire screener phase confirmed 349,664 qualifying farms that were added to the CML. A total of 282,901 names were confirmed as out of scope and were dropped from the list. The number of names returned as undeliverable as addressed was 92,203 and they were excluded from further census mailings. The remaining 435,212 did not respond and were mailed census forms although they were not added to the CML as active farms.

During the spring and summer of 2002, measures were taken to improve name and address quality. Checks were made to detect and remove duplication both within states and across states. List addresses were processed through the National Change of Address registry and the Locatable Address Conversion System to ensure they were correct and complete. Records on the list frame with missing or invalid phone numbers were matched against a nationally available telephone database to obtain as many phone numbers as possible.

Records requiring special handling for census data collection or for analysis and summarization were identified. These were mostly farms considered unique because of their size or because they produce specialty commodities.

The official Census Mail List was established on September 1, 2002. The list contained 2,841,788 records. These records can be broken down into 1,839,533 records that were thought to meet the NASS farm definition and 1,002,255 potential farm records.

CENSUS SAMPLE DESIGN

All name and address records on the final CML received a 2002 Census of Agriculture report form. Two different types of census report forms, sample and

nonsample, were used to collect data. Sections 1 through 16 and 22 through 25 of the sample form (long form) were identical to sections on the nonsample census form (short form). Sections 17 through 21 of the sample form contained additional questions on usage of fertilizers and chemicals, farm production expenditures, value of machinery and equipment, value of land and buildings, and hired workers. There were 12 regional versions of the nonsample form and 13 regional versions of the sample form with listings of crops varying by region.

The sample form was mailed to all mail list records in Alaska and Rhode Island and to a sample of records in other States selected from the final mail list. Mail list records were selected into the sample with certainty if they (1) were expected to have large total value of agricultural products sold or large acreage, (2) were in a county with less than 100 farms in 1997, or (3) had other special characteristics (e.g., abnormal farms such as institutional farms; experimental and research farms; Indian reservations; etc.). Mail list records in counties containing 100 to 199 farms in 1997 were systematically sampled at a rate of 1 in 2; counties containing 200 to 299 farms in 1997 were systematically sampled at a rate of 1 in 4; counties containing 300 to 399 farms in 1997 were systematically sampled at a rate of 1 in 6; and counties containing 400 or more farms in 1997 were systematically sampled at a rate of 1 in 8. The mail list records not chosen to receive the sample form received the nonsample census form. This differential sampling scheme was used to provide reliable data for the sample sections of the report form for all counties.

The regional report form versions and the sampling scheme were used to provide reliable data for a large number of items/commodities at the county level, while reducing response burden.

EDITING DATA AND IMPUTING FOR ITEM NONRESPONSE

The mailing label on all forms returned to the National Processing Center (NPC) were scanned using bar code readers to capture identifiers and for check-in purposes. Forms determined to represent qualifying, in-scope farms were submitted for imaging. A snapshot was taken of each page of every report form and optical mark recognition (OMR) and intelligent character recognition (ICR) techniques were used to capture reported data from the images. The ICR engine determined a confidence level for every cell read. Any cell with a confidence level below a prescribed value was referred to NPC staff to review and correct from the image, when necessary. The images and the captured data were transferred to NASS on a flow basis. Data collected by telephone were captured using computer assisted telephone interview software. Data entry procedures were developed for NASS field offices to input data from forms received too late to be imaged.

Captured data were processed through a format program. This program verified that record identifiers were valid and checked the basic integrity of the data fields. Rejected records were referred to NASS staff for correction. Accepted records were posted to the database.

All 2002 Census of Agriculture data were passed through a complex computer edit. Data were batched by state for submission to the computer edit. The edit first determined whether a reporting operation met the minimum criteria to be counted as a farm in the census. Operations failing to meet the minimum criteria were referred to NASS staff for verification. The edit examined each report for reasonableness and completeness and determined whether to accept, delete, impute (supply), or alter the reported value for each data record item.

Whenever possible, imputations, deletions, and changes made by the editing system were based on related data on the respondent's report form (deterministic imputation). For some items, such as operator characteristics, available data for that farm from the previous census were used. Values reported on previous NASS surveys were used, where applicable.

When these and similar methods were not available and values had to be supplied, the imputation process used information reported for another farm operation in the same state or in a neighboring state with characteristics similar to those of the farm operation with incomplete data. For example, a farm operation that reported acres of corn harvested, but did not report bushels of corn harvested, was assigned the same bushels of corn per acre harvested as that of another

farm from that region having similar characteristics and reporting an acceptable yield. The imputation for missing items in each section of the report form was conducted separately; thus, assigned values for one operation could come from more than one respondent.

Each execution of the computer edit consisted of records from only one state. Successfully edited records were made available as potential "donors," to supply values needed in subsequent imputations. These "clean" records were accumulated into pools of donors according to geographic location, so that each pool might be used during the computer edit of any reports from appropriate states. When imputation was required, a report's collective imputation needs for a section were used to identify a group of "matching" variables for the report which contained acceptable data relating to the missing items. For example, acres of corn harvested would be a matching variable for bushels of corn harvested, in consideration of the high correlation between the two items.

Similarity to the report being edited was evaluated for the matching variables for all farms in the appropriate donor pool. Values were imputed from the donor report considered most similar, referred to in this context as the "nearest neighbor" to the report being edited. Similarity between the edited record and a donor was calculated as the Euclidean distance between their selected matching variables. As part of the distance computation, the values of the matching variables were normalized to have the same variance within each donor pool. Latitude and longitude were consistently included in all imputation requests as matching variables, so that geographic proximity played a role in all donor selections.

Imputation conformed to logic provided by the complex edit. When appropriate, only donors able to contribute a nonzero imputed value were considered. For a farm reporting harvested corn acreage, for example, imputed bushels of corn harvested would be taken only from farms with harvested corn. In addition, imputed values were often adjusted. In some cases, acceptable data in another field of the edited report were used to establish a ratio between the edited report and the donor report. This proportion was applied to the imputed value as a scale factor. In the corn example, total bushels of corn from the donor would be scaled by the ratio of the acres of corn in the edited report to those in the donor report.

To maintain consistency with the complex edit, the imputed values in most sections of the report were tested to ensure they satisfied critical relationships among items within the section. If any of these constraints were not met, alternative donors were considered in order of their similarity to the edited report, until all the constraints for the module were satisfied.

In some cases, nearest-neighbor imputation was not possible. The requirement of a positive imputed value might rule out all available donors, resulting in an imputation failure. However, if some members of the donor pool were found to satisfy this requirement, then as many as 25 nearest neighbors were given further consideration. But if none of the candidate donors could provide qualifying data, the result was also noted as an imputation failure. Processing of records that encountered these imputation failures was suspended at the section where the failure occurred. These records were made available for analyst review and later reconsidered by the automated edit as a follow-up to corrective actions taken by the analyst.

The donor pool for each region was frequently updated with records from its area which had completed the editing process. As records were added to the donor pool, the records became available to donate values to incomplete reports subsequently edited for that region. Prior to editing, all donor pools were empty and no donors were available. Initial donor pools were created by giving special treatment to the first batches of data received from each state. Similar to the way that imputation failures were resolved through analyst review of the reports, early reports from initial batches were reviewed and adjusted manually by teams of analysts. This process was employed until each donor pool became self-sufficient in consistently providing imputed values for its region through the automated nearest-neighbor selection process.

To streamline editing once they had reached a mature stage in their growth, donor pools for some regions were not expanded in size beyond a chosen plateau. This provided assurance that computer edits would not exceed a reasonable processing time for nearest-neighbor searches. Although their size was limited, these donor pools did not become static. They were

regularly recreated with representative samples of all records available from their regions. Within a given region, all successfully edited long form records were included in their appropriate donor pool. Successfully edited short form records were ordered by farm size and sales volume for a given region, and then systematically sampled. Every "ith" record from the short-form list was joined to the complete list of long forms for its region to form a refreshed donor pool. The steady renewal of donor pools for regions with large numbers of records assured a more diverse selection of donors over time.

All records for which data were changed were resubmitted to the edit to verify an acceptable correction was made. Records with imputation failures were referred to an analyst for resolution. A data review screen presented the problematic data. The analyst could summon the image, the census mail list, or the historical data warehouse to help determine a suitable solution. Corrected data were posted and the record was re-edited.

The complex edit ensured the full internal consistency of the record. Analysts were provided an additional set of tools to review record-level data across farms. These examinations detected extreme outliers or unique data distribution patterns that were possibly a result of reporting, recording, or handling errors. Potential problems were researched and, when necessary, corrections were made and the record re-edited.

NONRESPONSE AND SAMPLE ESTIMATION

Statistical estimation procedures were used to account for whole farm nonresponse and sample data collection. The procedures for nonresponse were necessary because some farm operators did not respond to the census despite numerous attempts to contact them. Statistical estimates for long-form-only data items had to be calculated since, by design, the data were not collected from every farm.

Treatment of Farms Selected for the Screener Phase

Names selected to receive the screener form were those believed to have some likelihood of operating a farm, but for whom actual farm activity was unknown.

The screener phase and follow up strategies resulted in several possible outcomes depending on whether the screener name responded and was in or out of scope. Each of these outcomes was handled differently to adjust for nonresponse.

Names responding to the screener as out of scope (nonfarms) were excluded from the CML. If the respondent answered the screener as in scope, the respondent was added to the CML and received a census form. If this in-scope screener respondent answered the census form, the operation's report was eligible to be used to help account for nonrespondents to the census. If the in-scope screener respondent failed to respond to the census form, that operation's data were accounted for by census respondents.

Records for operation that did not respond to any of the three mailings of the screener were not considered to be part of the CML. Nevertheless, they were sent a census form. If the screener nonrespondent ultimately responded as an in-scope operation on the census, it was given a fixed nonresponse weight of 1 for census tabulations. If the screener nonrespondent failed to respond to the census form, the record was treated in summarization as if it never existed.

Whole Farm Nonresponse Estimation

Whole farm nonresponse to the census occurred when no data were received from an operation on the CML. If the record was deemed to represent either a large farm, as defined by the total value of production or acreage, or a unique farm operation, intensive telephone or personal followup was conducted during the census processing to obtain a response. If these attempts failed, the NASS survey database, the census historical database, or other more current sources were used to impute data for the record. These large and/or unique records were designated as must records and were assigned a fixed nonresponse weight of 1, meaning their data were not used for nonresponse adjustment. Screener respondents with reported sales above a certain state-determined level automatically became must records.

During mail list development, the State Statistical Offices (Field offices), in an effort to reduce respondent burden, identified operations that participated in multiple NASS surveys, and those that

had special reporting relationships with an enumerator. The records for these operations were "tagged". The Field offices assumed full responsibility for the data collection for any tagged operations, including imputation of data for them if a response was not obtained. Tagged records became must records. They had a nonresponse weight of 1 and the reports were not be used for nonresponse adjustments.

Whole farm nonresponse that occurred within the remaining universe of records, called non-musts, was accounted for by a statistical weighting procedure. All responding non-musts in a state were put into mutually exclusive weighting groups based on their size and county as recorded on the CML database. Statistical models were used to estimate the number of nonresponse farms that were in scope for each weighting group. The weights of the responding farms in each weighting group were increased to account for nonresponding farms in that group.

Throughout the data collection period, there were changes and additions to the CML. Records added after the initial CML was created on September 1, 2002 were designated as new adds and treated like screener nonrespondents and given a nonresponse weight of 1. New adds responding as in-scope records to the census were subsequently subtracted from the measurement of undercoverage. When a new add was linked to an operation originally on the CML, it was no longer considered a new add. New adds occurred any time after the CML creation and before final weighting in February, 2004.

Some operators were sent more than one census form. These operators were required to fill out a separate form for each operation. Also, an operator may have had an operation for which a census form was not received, but the existence of which was noted on the form of the known operation. That operator was sent a new census form or enumerated by telephone to obtain data for that previously unknown operation. If a response was obtained for the previously unknown operation, the nonresponse weight for the new record was set equal to the nonresponse weight for the original operation reporting its existence. If no response was obtained for the previously unknown operation, it was treated as out of scope.

Some large farms operating in more than one county were treated as distinct county-specific operations to more accurately allocate data to counties. Similarly, large farms operating in more than one state were treated as distinct state-specific operations. Split add records were created for these operations and they were assigned the same nonresponse weight as the original CML operation.

Controls were established that ensured the calculated nonresponse weight never exceeded 2. The nonresponse weights were systematically rounded to integers, with an integerized weight of either 1 or 2 assigned to each record. The integerization process eliminated any impact rounding has on census farm counts and totals in each county and in cross tabulations.

Tables A and C quantify the effect of the nonresponse estimation procedures on selected census data items. These tables contain percentages of the census aggregates that were contributed by nonresponse adjustments. As noted earlier, names included in the screener sample that never responded were treated as if they never existed. Any in-scope farm in this group was missed and, consequently, "attributed" to the coverage adjustment. This is shown in Table C. For selected items, estimates of what was attributed were reallocated to nonresponse to obtain "corrected" values, which appear in Table A. This was possible at the state level only. The differences between state-level nonresponse adjustment numbers in the first line of Table C and their counterparts in Table A represent the amount reallocated.

There was no such reallocation in Hawaii because records in that state were not adjusted to account for coverage errors. No tables appear for Alaska, because those state's records were not adjusted for nonresponse or coverage.

The estimates provided in Tables A and C do not reflect the effect of item nonresponse to individual census data items. The effect of this item nonresponse is discussed in the ''Nonmeasurable Census Error'' section.

Sample Estimation

Must records were all preselected to receive the census

long form. Non-must records were sampled to determine which would receive the long form and which the short form. All records in some small counties automatically received the census long form. However, these records were not necessarily must records. Nonresponse adjustment was allowed for the non-musts.

Weights applied to the items appearing on the long form only (Sections 17 through 21) were calculated by multiplying the farm's coverage-adjusted weight, which is described later, by the sample factor (e.g, 6 for a farm sampled with a 1-in-6 rate, 1 for a must). An adjustment was made that ensured the number of farms operating in a county as estimated from the sample matched the number estimated from the full census. Before computing published tabulations based on the sample, each record's sample weight was integerized to eliminate the impact of rounding on census farm counts and totals.

Operators with more than one operation were sampled as one record and received the same census form for each operation. Operations added after sampling were treated differently depending on whether or not the record was linked to a record on the original CML. Added operations which linked to a record on the original CML were mailed the same census form as the original CML operation. Added operations that were not linked to a record on the original CML were mailed the long form.

MEASURABLE CENSUS ERROR

The root mean squared error of an estimated data item from the census provides a measure of the error a field office associated with completing a census. It measures the variation in the value of that estimated data item based on all possible outcomes of the census collection, including variants as to who was on the census list, who returned a census form and who was selected to fill out the sample form.

Data items are classified as either complete count items or sample count items. Sample count items were collected only on the longer sample version of the census report form. Complete count items were collected from all respondents. Variability in the estimates of complete count items was due only to the nonresponse and coverage estimation adjustment

procedures. Variability in the estimates of sample count items was due to both the adjustment procedures and the census sample selection and estimation procedure. Therefore, variability in the sample count item estimates tends to be larger than the variability in the complete count item estimates.

Table B presents the fully adjusted total with the root mean squared error for selected items. The relative root mean squared error is obtained by dividing the root mean squared error by the value of the estimate multiplied by 100. The table also includes the percent contribution to the mean squared error (the square of the root mean squared error) from nonresponse adjustment and sampling and from coverage adjustment.

There is no Table B for Alaska. Mean squared errors in Hawaii displayed in Table B are entirely due to nonresponse adjustment.

Nonsampling error due to mail list incompleteness and duplication as well as misclassification of records on the mail list is called coverage error. The section titled "Classification Error" addresses attempts to assess, at least qualitatively, the impact of classification error on the census results.

NONMEASURABLE CENSUS ERROR

The accuracy of the census counts is affected jointly by the measurable errors described in the previous section
and by nonmeasurable errors (nonmeasurable in the sense of not being included in root mean squared error estimates). Extensive efforts were made to compile a complete and accurate mail list for the census, to design an understandable report form with instructions, and to minimize processing errors through the use of quality control measures. Despite these efforts, nonmeasurable errors are inevitable and arise from many sources, including respondent or enumerator error, incorrect data capture, editing, and imputing for missing data. These errors are discussed in this section.

Respondent and Enumerator Error

Incorrect or incomplete responses to the census report form or to the questions posed by an enumerator can

introduce error into the census data. To reduce reporting error, detailed instructions for completing the report form were provided to each respondent. Questions were phrased as clearly as possible based on previous tests of the report form. Computer-assisted telephone interviewing software included immediate integrity checks of recorded responses so suspect data could be verified or corrected. In addition, each respondent's answers were checked for completeness and consistency by the complex edit and imputation system.

Item Nonresponse

As information flowed from data collection to tabulation, various types of item nonresponses were identified on the census report forms. Nonresponse to particular questions on the form that logically should have been present created a type of nonsampling error in both complete count and sample count data. In this case, information from a similar farm was used to impute for these missing data items. The resulting data may have been biased if the characteristics of the nonreporting farms were different from those of reporting farms for those items. The section titled "Editing Data and Imputing for Item Nonresponse" provides a detailed explanation of item imputation procedures.

Processing Error

All phases of processing for each census report form were potential sources of nonsampling error. An automated check-in procedure recorded that the report had been returned and excluded it from further followup mailings. Approximately one-third of the mail returns were reviewed to resolve questions dealing with multiple reports, respondent remarks, or no reported data. The remaining mail returns (about two-thirds), along with some of the reviewed cases containing farm data, were batched and sent directly to imaging and data capture. Data were transmitted, formatted, and run through the complex edit and imputation system to ensure within record consistency. About one-fifth of all forms edited were clerically reviewed for inconsistencies, omissions, or questionable values. While reviewing these forms, staff determined if the action taken by the computer edit and imputation system was correct. Additional analysis tools were used to examine data across records for distributional irregularities and extreme values. Edited records were tabulated to the county level. Each county was reviewed and, when necessary, individual records were corrected prior to publication.

Developing accurate processing methods is complicated by the complex structure of agriculture. Among the complexities are the many places to be included, the variety of arrangements under which farms are operated, the continuing changes in the relationship of operators to the farm operated, the expiration of leases and the initiation or renewal of leases, the problem of obtaining a complete list of agriculture operations, the difficulty of contacting and identifying some types of contractor/contractee relationships, the operator's absence from the farm during the data collection period, and the operator's opinion that part or all of the operation does not qualify and should not be included in the census. During data collection and processing of the census, all operations underwent a number of quality control checks to ensure results were as accurate as possible.

COVERAGE ADJUSTMENT

Although much effort was expended making the CML as complete as possible, the coverage of farms was not complete. NASS's goal was to produce agricultural census totals for publication that were fully adjusted for list undercoverage at the county level. To this end, estimates of the undercoverage for a specified set of farm characteristics, called calibration variables, were computed using an area-frame sample. After initial weights were assigned to census respondents to account for nonresponse, these weights were further adjusted to compensate for estimated state level undercoverage for each of the calibration variables based on the area frame sample. Since each farm with census data was given a fully-adjusted weight by this process, county level totals could be generated for every census variable not just the calibration variables. The section titled "Calibration Algorithm" provides a list of the area frame based calibration variables.

To further improve coverage adjustment, a second set of targets and ranges were added to the calibration effort. These were well established commodity totals for which excellent check data were available for validation. The introduction of these commodity target strengthened the overall coverage adjustment

process by limiting the possible adjustments produced by the area frame based targets to ensure major commodity totals remained within reasonable bounds of established benchmarks.

Most targets were determined at the state level. The one exception was the New England states - Connecticut, Maine, Massachusetts, New Hampshire, Rhode Island, and Vermont - which were combined into one "calibration region". In what follows, "state" refers to the calibration region for New England.

Measuring Mail List Undercoverage

Census mail list undercoverage was measured using an independent survey of land segments selected from the NASS area frame. The NASS area frame covers all land in the United States and includes all farms. Each June, NASS conducts a survey in which area frame segments are enumerated for agricultural activity. The sampled segments are allocated to provide accurate measures of acres planted to widely grown crops and inventories of hogs and cattle.

The 2002 June Area Survey (JAS) was supplemented by the 2002 Agricultural Coverage Evaluation Survey (ACES) to better estimate CML incompleteness. The ACES used a sample of segments allocated in a way that, when pooled with the JAS, accurate measures of number of farms and land utilization could be obtained. Enumerators visited all segments, identified all farms operating land in each segment, and obtained basic data about those farms.

The names and addresses collected in the 2002 JAS and 2002 ACES were matched to the census mail list. Farms that did not match were recontacted after the census mailout to confirm that they did not receive a census form. Farms that had not received a census form represented the farms not on the mail list (NML). Those who received a census form had been erroneously classified as NML and were removed.

The percentage of farms missed in the census varied considerably by State. In general, farms not on the mail list tended to be small in acreage, production, and sales of agricultural products. Farm operations were missed for various reasons, including the possibility that the operation started after the mail list was developed, the operation was so small as not to appear

in any agriculture-related source lists, or the operation was falsely classified as a nonfarm prior to mailout.

Determining Targets to Correct for Undercoverage

The 2002 June Agricultural Survey consisted of 11,075 land segments and the Agricultural Coverage Evaluation Survey (ACES) added 2,400 segments. Data values a field office associated with NML tracts were used to estimate the state-level undercoverage of the CML for the first set of calibration variables. The state-level totals for these variables were then summed to yield national totals.

The national NML estimate for the number of farms was used directly in determining calibration targets (CML
+ NML). State-level farm-count estimates based on the NML sometimes had unacceptably high standard errors, as well as apparent systematic biases. These estimates were smoothed across states based on separate NASS surveys and previous analysis.

Other calibration targets were derived from the NML-estimated fractions of farms of certain types (e.g., in a particular sales class or with a primary operator of a particular race). Most of these had unacceptably high state-level standard errors. As a result, more reliable national level NML estimates were used to smooth state estimates. The smoothed state NML-estimated fraction was computed by taking a weighted average of the actual state estimate and a prediction for the state based on national and state level numbers (e.g., the number of NML farms in the state, the fraction of farms with black owners on the state's CML, and the national relative difference between the fraction of black owners on the NML and CML). The weighting factor was chosen to approximately minimize mean squared error under a random effects model. The smoothed NML-estimated fractions were multiplied by the corresponding smoothed NML farm-count estimates described above and added to corresponding CML estimates to obtain coverage-adjusted state-level totals, which served as calibration targets.

Tolerance Ranges

Although full calibration would assure that the weighted total among census respondents equaled its for each calibration variable in either set, it was not always possible to calibrate to such a large number of target values while keeping all farm weights within a reasonable range (for example, the weight for any farm cannot be less than one). Because of this and because calibration targets are estimates themselves subject to uncertainty, NASS allowed some tolerance in the determination of coverage-adjusted weights. Rather than forcing the total for each calibration variable computed using the coverage-adjusted weights to equal a specific amount, NASS allowed the estimated total to fall within a tolerance range. This tolerance strategy sometimes made it possible for the calibration algorithm to produce a set of satisfactory coverage-adjusted weights that it would not have otherwise.

Ranges for the first set of calibration variables used to adjust for undercoverage were determined differently from the second set used to adjust for measurement error. The number of farms had no tolerance range. The tolerance range for every other variable in the first set was the estimated state total for the variable (CML + NML) plus or minus one-half of one estimated standard error. This choice limited the cumulative deviation from the estimated total for a variable when state-level totals were combined to create a US-level total.

The state-level tolerance ranges for commodity targets were provided by commodity specialists in NASS's Statistics Division. These ranges did not have to be symmetric around the target value.

Calibration Algorithm

Coverage adjusted weights were obtained by an algorithm based on the restricted regression algorithm referred to by Singh and Mohl (1996) as the Linear Truncated Method. Coverage adjustments began with the nonresponse weights before integerization. The final coverage-adjusted (nonsample) weights were restricted to the interval [1,6).

The calibration variables were based on the following reported items:

1. Total value of production and government payments.

0	$5,000 - $24,999
$1 - $999	$25,000 - $99,999
$1,000 - $2,499	$100,000 - $499,999
$2,500 - $4,999	$500,000 and above

2. Age of principal operator.

Less than 25 years old
25 - 34
35 - 44
45 - 54
55 and older

3. Sex of principal operator.
Female

4. Race of principal operator (selected categories).
Black
American Indian, Asian, and Other

5. Principal operators of Spanish, Hispanic, or Latino origin.

6. Number of farms and land in farms.

7. Selected types of farms by commodity produced.

All cattle farms

Dairy farms
Sheep/goat farms

Nursery/horticulture farms
Hog/pig farms
Fruit/nut/berry farms
Vegetable farms
Tobacco farms
Horse/Equine farms
Poultry farms

8. Various commodity acreage and production statistics (Varies by state).

Corn acres harvested
Soybean acres harvested
Wheat acres harvested
Potato acres harvested
Rice acres harvested

Sugarcane acres harvested
Hay acres harvested
Apples acres harvested
Total orange acres
Grape acres harvested
Cotton bales produced
Beef cow inventory
Lettuce acres harvested
Tomatoes acres harvested
Tobacco acres harvested
Sugarbeet acres harvested
Cattle on feed inventory
Total cattle inventory
Total hog/pig inventory
Dairy cow inventory
Broiler production
Layer inventory
Durum wheat acres harvested (North Dakota)
Other spring wheat acres harvested (North Dakota)
Alfalfa acres harvested (South Dakota)

Integerization and Sample Weights

Coverage-adjusted weights were integerized to eliminate the need for rounding estimated counts computed with coverage-adjusted weights. The integerization process was designed to minimize county-level impact on the nonresponse and coverage adjustment of number of farms and total land in farms.

Sample weights were computed by multiplying coverage-adjusted weights before integerization with the appropriate sampling factors and adjusting the results to add up to matched census counts as described previously. Sample weights were then integerized for analytical purposes.

Measuring the Amount of Coverage Adjustment

Tables A and C display the proportions of selected census data items that are due to nonresponse and coverage adjustments. The section of this appendix on whole farm nonresponse adjustment explained how the nonresponse adjustment values were determined. The coverage adjustment values account for the rest of the differences between the weighted and unweighted totals for these data items. Some estimated coverage adjustments are negative. The use of commodity targets in calibration indirectly exposed some

duplication on the census list resulting in negative coverage adjustments.

CLASSIFICATION ERROR STUDY

The 2002 Classification Error Study (CES) was conducted for the entire U.S. to study the potential impact of classification error on the census results. The study used the 2002 June Agricultural Survey (JAS) and 2002 Agricultural Coverage Evaluation Survey (ACES) to study farms incorrectly classified as nonfarms (undercount), nonfarms incorrectly classified as farms (overcount), and duplication of farms (overcount) in the 2002 Census of Agriculture. The CES was not intended to adjust census farm counts, but rather, to evaluate procedures and to identify potential improvements in list building, data collection, and other activities in preparation for future censuses.

For the evaluation, additional name, address, and telephone information were collected on both the JAS and ACES by adding the following three questions:

1. During the past two years, has the operator received mail for this operation at any address other than the one shown on the face page?

2. Excluding partners and landlords, were any other names a field office associated with this operation in the past year? (For example, other business names, spouses names, etc).

3. Is any of the land inside the blue tract boundary rented from others? (Include land for which you paid cash rent, land used rent free, or land rented on shares).

The CES consisted of a two phase review process. The initial phase, Review of Possible Matches, used Probabilistic Record Linkage (PRL) to match the additional information collected on the area surveys to the name and addresses on the 2002 Census Mail List (CML) including late adds. PRL is a technique used to identify records that are believed to correspond to a CML record. Records were brought together into link groups, with each link group consisting of all records that possibly represented the same operation. Each link group was classified into one of three distinct types: matches, possible matches and nonmatches.

The nonmatches were represented in estimation as part of the undercoverage measure. The CES was primarily concerned with the matches and possible matches. Each State office reviewed the possible matches and determined match or nonmatch status.

Upon completion of the PRL review, there was a Farm Classification Resolution review by state offices of two additional sets of records. The first of these was comprised of area records matching two or more census records. Reviewing these records helped identify duplication on the CML. The second set consisted of groups of records (area and census) within which the reported acreage differed by more than 25 percent. A data analysis application was developed for analysts to review of the cases in the second phase. Upon completion of both phases, data were compiled to estimate undercount, overcount and duplication.

The analysis of these data will provide insight into census processes used to accurately determine farm status and identify duplication. Any weaknesses identified in the findings will be addressed for future censuses.

Table A. Summary of State Nonresponse and Coverage Adjustments: 2002

[For meaning of abbreviations and symbols, see introductory text]

Item	Total	Percent from nonresponse adjustment, corrected	Percent from coverage adjustment, corrected
Farms .... number	77,797	12.4	18.3
Land in farms .... acres	14,583,435	10.5	6.9
Farms by size:			
1 to 9 .... farms	7,471	11.5	36.6
acres	36,872	11.5	36.8
10 to 49 .... farms	23,261	11.2	29.2
acres	605,295	11.3	27.3
50 to 179 .... farms	27,427	13.0	14.6
acres	2,736,833	13.2	13.2
180 to 499 .... farms	12,615	15.5	3.6
acres	3,663,741	15.6	3.0
500 to 999 .... farms	4,309	14.3	1.7
acres	2,967,746	14.0	1.8
1,000 to 1,999 .... farms	2,107	3.8	6.9
acres	2,826,809	3.6	7.2
2,000 or more .... farms	607	0.8	6.6
acres	1,746,139	0.7	5.5
Market value of agricultural products sold $1,000	4,263,549	7.0	3.7
Farms by value of sales:			
Less than $1,000 .... farms	18,939	11.1	33.1
$1,000	2,474	14.3	19.3
$1,000 to $2,499 .... farms	9,979	10.4	33.9
$1,000	16,517	10.2	33.5
$2,500 to $4,999 .... farms	9,022	12.9	17.9
$1,000	32,110	12.9	17.6
$5,000 to $9,999 .... farms	8,718	13.5	9.9
$1,000	61,886	13.5	9.6
$10,000 to $19,999 .... farms	8,161	13.9	9.6
$1,000	115,285	14.0	9.4
$20,000 to $24,999 .... farms	2,482	14.3	9.4
1,000	55,021	14.3	9.3
$25,000 to $39,999 .... farms	4,431	15.8	5.0
$1,000	139,281	15.8	4.9
$40,000 to $49,999 .... farms	1,900	15.6	4.4
$1,000	84,619	15.7	4.3
$50,000 to $99,999 .... farms	5,297	16.2	6.2
$1,000	373,061	16.3	6.0
$100,000 to $249,999 .... farms	5,384	13.3	6.0
$1,000	845,567	12.7	6.1
$250,000 to $499,999 .... farms	2,288	5.4	5.7
$1,000	780,893	5.0	5.3
$500,000 to $999,999 .... farms	767	2.7	-0.8
$1,000	511,592	2.6	-0.7
$1,000,000 or more .... farms	429	1.2	0.7
$1,000	1,245,243	0.5	0.3
Farms by type of organization:			
Family or individual .... farms	70,890	12.6	18.9
acres	11,553,861	11.6	7.5
Partnership .... farms	4,549	12.5	10.2
acres	2,011,507	7.7	3.0
Corporation:			
Family held .... farms	1,689	8.3	15.0
acres	864,583	3.7	5.8
Other than family held .... farms	154	6.5	24.7
acres	51,433	0.6	20.2
Other - cooperative, estate or trust, institutional, etc .... farms	515	9.1	17.9
acres	102,051	7.3	8.7
Tenure:			
Full owners .... farms	52,168	12.0	22.6
acres	4,763,213	12.0	12.6

Item	Total	Percent from nonresponse adjustment, corrected	Percent from coverage adjustment, corrected
Tenure - Con.			
Part owners .... farms	20,863	13.0	8.3
acres	8,742,151	9.5	3.6
Tenants .... farms	4,766	15.2	15.0
acres	1,078,071	11.9	7.4
Principal Operator Characteristics:			
Sex of operator:			
Male .... farms	70,137	12.5	17.3
acres	13,962,956	10.4	6.7
Female .... farms	7,660	11.8	27.1
acres	620,479	13.1	11.3
Primary occupation:			
Farming .... farms	43,488	12.5	14.5
Other .... farms	34,309	12.3	23.0
Spanish, Hispanic, or Latino origin (see text) .... farms	804	6.6	59.5
acres	124,008	4.6	54.4
Race:			
White .... farms	77,213	12.5	18.2
acres	14,513,126	10.5	6.8
Black or African American .... farms	168	14.3	28.6
acres	(D)	(D)	(D)
American Indian or Alaska Native .... farms	192	9.9	40.6
acres	(D)	(D)	(D)
Native Hawaiian or Other Pacific Islander .... farms	8	0.0	37.5
acres	1,499	0.0	42.9
Asian .... farms	32	3.1	40.6
acres	2,728	2.4	37.1
More than one race reported ..... farms	184	10.9	36.4
acres	(D)	(D)	(D)
Age group and primary occupation:			
Reporting primary occupation as farming by age group:			
Under 25 years .... farms	443	11.1	45.8
25 to 34 years .... farms	2,459	14.7	23.2
35 to 44 years .... farms	7,726	12.2	20.4
45 to 54 years .... farms	10,007	12.3	14.1
55 to 64 years .... farms	9,851	12.1	12.3
65 years and over .... farms	13,002	12.9	10.4
Reporting primary occupation as other than farming by age group:			
Under 25 years .... farms	303	11.6	50.8
25 to 34 years .... farms	2,165	13.7	30.3
35 to 44 years .... farms	7,819	13.0	26.9
45 to 54 years .... farms	11,442	12.5	21.9
55 to 64 years .... farms	7,932	11.5	20.7
65 years and over .... farms	4,648	11.8	18.0
All operators by age group [1]:			
Under 25 years .... farms	2,918	12.0	29.9
25 to 34 years .... farms	8,655	13.7	22.7
35 to 44 years .... farms	24,467	12.4	22.3
45 to 54 years .... farms	30,505	12.3	18.8
55 to 64 years .... farms	23,696	11.7	16.7
65 to 74 years .... farms	13,954	12.0	13.2
75 years and over .... farms	7,222	13.5	11.2

[1] Data were collected for a maximum of three operators per farm.

Item	Total	Root mean squared error (RMSE)	Relative RMSE (percent)	Nonresponse and sampling contribution to MSE (percent)	Coverage adjustment to MSE (percent)
Farms .... number	77,797	2,217	2.9	(Z)	100.0
Land in farms .... acres	14,583,435	459,728	3.2	0.2	99.8
Farms by size:					
1 to 9 .... farms	7,471	320	4.3	1.3	98.7
acres	36,872	1,524	4.1	1.8	98.2
10 to 49 .... farms	23,261	734	3.2	0.6	99.4
acres	605,295	18,929	3.1	0.7	99.3
50 to 179 .... farms	27,427	857	3.1	0.5	99.5
acres	2,736,833	86,883	3.2	0.5	99.5
180 to 499 .... farms	12,615	458	3.6	0.9	99.1
acres	3,663,741	134,644	3.7	0.9	99.1
500 to 999 .... farms	4,309	180	4.2	2.0	98.0
acres	2,967,746	125,077	4.2	2.0	98.0
1,000 to 1,999 .... farms	2,107	95	4.5	1.0	99.0
acres	2,826,809	127,278	4.5	0.9	99.1
2,000 or more .... farms	607	19	3.2	1.3	98.7
acres	1,746,139	47,494	2.7	0.9	99.1
Market value of agricultural products sold $1,000	4,263,549	118,838	2.8	0.2	99.8
Farms by value of sales:					
Less than $1,000 .... farms	18,939	1,189	6.3	0.1	99.9
$1,000	2,474	220	8.9	0.4	99.6
$1,000 to $2,499 .... farms	9,979	668	6.7	0.2	99.8
$1,000	16,517	1,103	6.7	0.2	99.8
$2,500 to $4,999 .... farms	9,022	509	5.6	0.4	99.6
$1,000	32,110	1,803	5.6	0.4	99.6
$5,000 to $9,999 .... farms	8,718	427	4.9	0.6	99.4
$1,000	61,886	3,036	4.9	0.7	99.3
$10,000 to $19,999 .... farms	8,161	396	4.8	0.8	99.2
$1,000	115,285	5,590	4.8	0.8	99.2
$20,000 to $24,999 .... farms	2,482	114	4.6	3.5	96.5
1,000	55,021	2,529	4.6	3.6	96.4
$25,000 to $39,999 .... farms	4,431	218	4.9	1.6	98.4
$1,000	139,281	6,871	4.9	1.6	98.4
$40,000 to $49,999 .... farms	1,900	99	5.2	3.7	96.3
$1,000	84,619	4,419	5.2	3.7	96.3
$50,000 to $99,999 .... farms	5,297	243	4.6	1.5	98.5
$1,000	373,061	16,970	4.5	1.6	98.4
$100,000 to $249,999 .... farms	5,384	247	4.6	1.1	98.9
$1,000	845,567	38,290	4.5	1.1	98.9
$250,000 to $499,999 .... farms	2,288	84	3.7	1.9	98.1
$1,000	780,893	28,057	3.6	1.8	98.2
$500,000 to $999,999 .... farms	767	39	5.1	1.1	98.9
$1,000	511,592	25,182	4.9	1.1	98.9
$1,000,000 or more .... farms	429	15	3.4	1.1	98.9
$1,000	1,245,243	24,912	2.0	1.2	98.8
Farms by type of organization:					
Family or individual .... farms	70,890	2,031	2.9	0.1	99.9
acres	11,553,861	377,397	3.3	0.3	99.7
Partnership .... farms	4,549	146	3.2	3.5	96.5
acres	2,011,507	61,584	3.1	2.3	97.7
Corporation:					
Family held .... farms	1,689	72	4.3	5.0	95.0
acres	864,583	31,698	3.7	2.8	97.2
Other than family held .... farms	154	13	8.8	16.7	83.3
acres	51,433	7,902	15.4	14.6	85.4
Other - cooperative, estate or trust, institutional, etc .... farms	515	25	4.9	14.2	85.8
acres	102,051	5,398	5.3	12.6	87.4
Tenure:					
Full owners .... farms	52,168	1,522	2.9	0.2	99.8
acres	4,763,213	142,040	3.0	0.8	99.2
Part owners .... farms	20,863	685	3.3	0.5	99.5
acres	8,742,151	302,002	3.5	0.4	99.6
Tenants .... farms	4,766	193	4.1	2.4	97.6
acres	1,078,071	49,850	4.6	2.7	97.3
Principal Operator Characteristics:					
Sex of operator:					
Male .... farms	70,137	2,054	2.9	0.1	99.9
acres	13,962,956	443,725	3.2	0.2	99.8
Female .... farms	7,660	514	6.7	0.3	99.7
acres	620,479	54,138	8.7	1.1	98.9
Primary occupation:					
Farming .... farms	43,488	1,274	2.9	0.3	99.7
Other .... farms	34,309	1,020	3.0	0.4	99.6
Spanish, Hispanic, or Latino origin (see text) .... farms	804	197	24.5	0.1	99.9
acres	124,008	35,059	28.3	0.9	99.1
Race:					
White .... farms	77,213	2,203	2.9	(Z)	100.0
acres	14,513,126	458,108	3.2	0.2	99.8
Black or African American .... farms	168	49	29.2	0.7	99.3
acres	(D)	(D)	(D)	(D)	(D)
American Indian or Alaska Native .... farms	192	43	22.3	2.1	97.9
acres	(D)	(D)	(D)	(D)	(D)
Native Hawaiian or Other Pacific Islander .... farms	8	3	35.8	16.9	83.1
acres	1,499	717	47.8	19.6	80.4

See footnote(s) at end of table.   --continued

[For meaning of abbreviations and symbols, see introductory text]

Item	Total	Root mean squared error (RMSE)	Relative RMSE (percent)	Nonresponse and sampling contribution to MSE (percent)	Coverage adjustment to MSE (percent)
Principal Operator Characteristics - Con.					
Race - Con.					
Asian . . . . farms	32	10	30.7	7.8	92.2
acres	2,728	985	36.1	12.8	87.2
More than one race reported . . . . farms	184	39	21.2	2.2	97.8
acres	(D)	(D)	(D)	(D)	(D)
Age group and primary occupation:					
Reporting primary occupation as farming by age group:					
Under 25 years . . . . farms	443	127	28.6	0.6	99.4
25 to 34 years . . . . farms	2,459	269	10.9	0.6	99.4
35 to 44 years . . . . farms	7,726	417	5.4	0.7	99.3
45 to 54 years . . . . farms	10,007	462	4.6	0.6	99.4
55 to 64 years . . . . farms	9,851	359	3.6	1.2	98.8
65 years and over . . . . farms	13,002	475	3.7	0.8	99.2
Reporting primary occupation as other than farming by age group:					
Under 25 years . . . . farms	303	76	25.2	1.2	98.8
25 to 34 years . . . . farms	2,165	217	10.0	0.7	99.3
35 to 44 years . . . . farms	7,819	410	5.2	0.7	99.3
45 to 54 years . . . . farms	11,442	524	4.6	0.6	99.4
55 to 64 years . . . . farms	7,932	283	3.6	1.6	98.4
65 years and over . . . . farms	4,648	181	3.9	2.5	97.5
All operators by age group [1]:					
Under 25 years . . . . farms	2,918	248	8.5	1.0	99.0
25 to 34 years . . . . farms	8,655	612	7.1	0.4	99.6
35 to 44 years . . . . farms	24,467	1,066	4.4	0.3	99.7
45 to 54 years . . . . farms	30,505	1,227	4.0	0.3	99.7
55 to 64 years . . . . farms	23,696	797	3.4	0.6	99.4
65 to 74 years . . . . farms	13,954	488	3.5	1.0	99.0
75 years and over . . . . farms	7,222	268	3.7	1.8	98.2
Net cash farm income of operation (see text) [2]:					
Farms with gains of [3] -					
Less than $1,000 . . . . farms	4,450	311	7.0	51.2	48.8
$1,000	2,119	167	7.9	53.2	46.8
$1,000 to $4,999 . . . . farms	8,930	469	5.3	36.6	63.4
$1,000	23,473	1,309	5.6	38.2	61.8
$5,000 to $9,999 . . . . farms	4,717	309	6.6	45.0	55.0
$1,000	34,031	2,246	6.6	45.3	54.7
$10,000 to $24,999 . . . . farms	5,932	346	5.8	39.4	60.6
$1,000	97,555	5,833	6.0	40.0	60.0
$25,000 to $49,999 . . . . farms	3,605	247	6.9	44.6	55.4
$1,000	127,215	8,794	6.9	46.0	54.0
$50,000 or more . . . . farms	5,203	255	4.9	31.6	68.4
$1,000	957,338	35,970	3.8	26.6	73.4
Farms with losses of -					
Less than $1,000 . . . . farms	5,529	359	6.5	51.1	48.9
$1,000	2,858	213	7.4	53.3	46.7
$1,000 to $4,999 . . . . farms	17,753	749	4.2	28.3	71.7
$1,000	48,233	2,142	4.4	32.4	67.6
$5,000 to $9,999 . . . . farms	9,655	525	5.4	41.1	58.9
$1,000	69,092	3,832	5.5	41.5	58.5
$10,000 to $24,999 . . . . farms	7,345	414	5.6	47.4	52.6
$1,000	113,909	6,573	5.8	48.0	52.0
$25,000 to $49,999 . . . . farms	2,829	238	8.4	58.9	41.1
$1,000	94,788	8,008	8.4	58.0	42.0
$50,000 or more . . . . farms	1,840	154	8.3	56.1	43.9
$1,000	218,308	17,202	7.9	63.7	36.3

[1] Data were collected for a maximum of three operators per farm.
[2] Data are based on a sample of farms.
[3] Farms with zero net cash income are included as farms with gains of less than $1,000.

Table C. Summary of Nonresponse and Coverage Adjustments by County

[For meaning of abbreviations and symbols, see introductory text]

Geographic area	All farms Total (number)	Nonresponse adjustment (percent)	Coverage adjustment (percent)	Land in farms Total (acres)	Nonresponse adjustment (percent)	Coverage adjustment (percent)	Sales Total ($1,000)	Nonresponse adjustment (percent)	Coverage adjustment (percent)
STATE TOTAL									
Ohio	77,797	9.3	21.4	14,583,435	8.9	8.5	4,263,549	6.0	4.8
COUNTIES									
Adams	1,320	11.5	17.4	198,277	12.3	11.1	20,461	10.7	7.1
Allen	968	10.3	13.0	188,150	11.1	0.1	41,264	9.8	-2.2
Ashland	1,089	8.8	20.5	161,100	9.4	8.8	50,232	5.9	7.9
Ashtabula	1,283	9.8	26.6	170,424	11.6	13.7	39,934	14.9	4.3
Athens	673	7.6	25.7	104,816	8.8	17.7	7,959	13.5	7.6
Auglaize	1,020	8.5	14.0	217,916	8.3	3.8	70,425	7.1	2.5
Belmont	753	10.2	21.4	141,908	9.7	9.3	14,776	9.5	14.3
Brown	1,400	10.9	19.1	220,729	11.1	15.2	31,178	12.4	11.7
Butler	1,060	10.4	26.5	138,044	13.1	6.8	35,253	12.1	1.2
Carroll	749	9.6	20.7	123,506	10.5	10.1	23,283	10.9	11.9
Champaign	937	11.1	19.0	207,554	10.4	-1.6	50,447	7.0	0.2
Clark	756	10.1	20.5	165,366	8.7	0.2	70,910	4.7	-0.6
Clermont	973	8.4	32.6	116,026	6.6	25.0	18,158	7.2	18.4
Clinton	811	11.5	19.4	238,805	9.7	11.7	52,864	10.0	6.9
Columbiana	1,184	11.4	25.8	136,080	12.4	11.8	43,559	13.5	5.1
Coshocton	1,043	7.6	22.0	179,643	7.2	11.4	34,557	4.9	4.5
Crawford	693	9.4	13.3	234,204	6.8	2.1	68,855	4.8	3.1
Cuyahoga	159	6.3	34.0	4,086	6.2	25.3	18,621	1.5	4.2
Darke	1,764	6.8	18.6	339,055	6.9	3.6	304,206	2.2	1.2
Defiance	982	9.8	14.0	208,994	9.3	7.0	43,203	8.0	3.7
Delaware	785	7.5	28.9	162,554	5.7	14.3	50,452	3.9	6.3
Erie	392	8.2	17.1	94,681	6.8	5.7	32,607	6.1	1.1
Fairfield	1,173	10.4	21.1	196,128	12.0	3.2	41,078	10.6	0.8
Fayette	480	11.5	14.4	203,212	7.3	3.3	46,331	7.3	0.6
Franklin	561	9.4	32.1	81,593	10.1	12.9	31,502	6.1	4.9
Fulton	783	10.1	12.3	197,410	9.2	-1.5	70,109	7.6	-0.6
Gallia	936	10.4	24.6	117,944	11.1	11.7	14,294	10.2	7.9
Geauga	975	8.0	34.5	66,474	8.6	23.2	22,735	4.9	14.7
Greene	819	8.9	20.9	168,568	7.9	1.1	50,558	6.5	-0.2
Guernsey	910	8.9	18.7	137,443	9.6	8.5	12,855	11.1	9.0
Hamilton	399	6.3	33.1	29,520	6.0	20.1	23,635	3.7	8.8
Hancock	976	12.3	15.0	262,095	11.4	2.2	46,183	9.2	2.7
Hardin	842	12.4	14.4	246,393	11.0	4.1	95,549	4.3	1.4
Harrison	450	8.9	20.9	138,423	6.6	23.9	16,830	3.3	34.5
Henry	844	9.1	9.4	236,273	9.1	-1.8	58,918	8.3	-1.7
Highland	1,381	11.5	17.2	273,263	11.9	8.8	43,585	12.1	10.2
Hocking	434	7.8	26.7	49,866	9.5	16.5	3,554	6.3	18.7
Holmes	1,809	8.1	29.0	206,603	7.4	17.7	97,044	6.2	14.9
Huron	865	8.6	20.2	228,346	6.7	6.1	66,397	4.8	4.3
Jackson	458	9.6	20.3	73,800	11.4	11.1	6,363	17.4	12.4
Jefferson	461	8.2	21.0	67,231	11.0	8.9	6,765	7.9	17.2
Knox	1,258	10.9	21.5	209,067	12.6	3.2	54,818	10.5	1.7
Lake	333	8.1	28.8	19,785	6.7	18.9	72,492	0.8	2.6
Lawrence	644	7.8	29.2	65,326	9.6	15.8	3,731	10.4	1.7
Licking	1,482	9.9	24.9	237,285	8.5	7.7	105,723	3.7	2.0
Logan	1,055	11.4	20.5	225,093	10.1	7.0	48,970	7.5	4.0
Lorain	975	8.2	28.6	161,918	7.6	24.0	97,787	3.7	6.8
Lucas	405	7.7	23.7	77,823	6.9	4.6	41,452	2.8	6.0
Madison	730	11.2	19.5	245,886	8.8	3.7	60,771	6.8	3.8
Mahoning	652	9.7	27.1	76,543	9.7	13.2	27,501	8.4	8.5
Marion	520	13.1	13.7	205,605	8.4	8.5	47,603	6.2	6.9
Medina	1,188	9.7	32.0	122,682	9.3	20.9	40,474	6.1	11.8
Meigs	552	10.5	18.7	90,362	11.0	13.7	19,133	6.0	5.2
Mercer	1,268	7.3	14.2	268,569	6.0	4.4	277,372	2.1	1.9
Miami	1,071	9.8	20.4	184,028	11.9	-2.2	39,809	10.5	-3.7
Monroe	654	7.2	17.7	107,198	8.1	8.2	7,127	7.9	9.7
Montgomery	832	8.9	26.0	101,912	9.9	5.7	32,940	5.9	2.4
Morgan	508	10.2	12.2	100,198	11.1	0.6	9,293	6.5	2.9
Morrow	863	9.3	23.9	179,051	7.2	9.4	35,747	6.6	5.9
Muskingum	1,222	7.1	20.6	193,175	8.0	12.1	25,614	11.7	2.5
Noble	602	8.6	20.1	106,957	9.0	11.9	5,270	11.3	13.2
Ottawa	517	8.5	21.5	114,430	7.1	19.0	23,907	5.3	13.5
Paulding	651	9.5	18.6	238,497	6.7	13.8	66,673	2.9	5.7
Perry	639	10.0	21.0	91,907	10.8	6.6	14,235	12.9	2.9
Pickaway	791	9.2	19.0	275,029	6.5	3.4	58,583	6.7	1.3
Pike	505	10.5	19.2	83,602	10.5	8.8	8,289	12.9	2.3
Portage	962	7.2	31.0	96,874	6.8	14.8	24,695	4.7	11.6
Preble	1,065	8.5	21.8	198,048	7.2	6.3	58,160	5.9	1.8
Putnam	1,348	7.6	15.9	331,517	6.7	13.5	88,326	4.9	8.7
Richland	1,086	9.9	22.2	158,653	9.8	10.3	46,354	10.7	12.1
Ross	952	8.8	18.7	246,690	7.4	4.4	37,468	6.4	0.8
Sandusky	802	9.7	17.0	196,152	8.8	3.4	51,045	7.4	3.0
Scioto	709	11.3	22.4	96,449	12.9	9.4	14,867	8.1	3.3
Seneca	1,185	9.9	14.6	280,449	9.8	4.8	55,599	8.5	3.2
Shelby	1,022	8.0	16.5	207,329	8.5	6.6	65,253	6.0	6.1
Stark	1,337	8.0	27.1	145,163	8.5	10.7	69,046	5.1	5.3
Summit	377	6.9	36.9	21,117	7.6	27.4	11,044	7.1	17.6
Trumbull	1,016	8.2	28.8	125,962	8.4	16.1	30,568	10.5	11.0
Tuscarawas	1,076	8.5	25.7	159,665	8.8	13.7	52,072	4.1	10.1
Union	1,021	9.7	24.0	256,024	5.8	20.6	88,770	2.9	10.1
Van Wert	681	8.1	12.9	250,224	7.0	8.4	59,057	6.9	5.2
Vinton	237	8.9	23.2	43,651	7.5	20.2	3,793	4.3	22.8
Warren	1,036	7.7	32.1	126,168	7.9	16.7	29,619	5.3	11.9
Washington	952	10.2	18.5	141,455	11.8	10.4	18,450	15.3	1.7
Wayne	1,894	7.6	25.4	267,169	6.9	13.7	158,638	3.9	9.5
Williams	1,099	9.0	15.3	213,265	9.4	5.5	44,606	6.7	4.6
Wood	1,066	11.4	13.9	305,834	10.4	5.3	81,177	7.9	3.8
Wyandot	607	9.2	13.5	201,146	7.3	2.0	72,144	2.6	0.8